

The Exposure Pyramid Framework and its Application to a Cross- Sectional Study of Lung Function

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Outline

- Introduction
 - the Exposure Pyramid Framework
 - Case study:
Cross-sectional study of exposure to indoor air pollution (IAP) and lung function in India
 - Conclusions
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Why is IAP from solid fuels problematic?

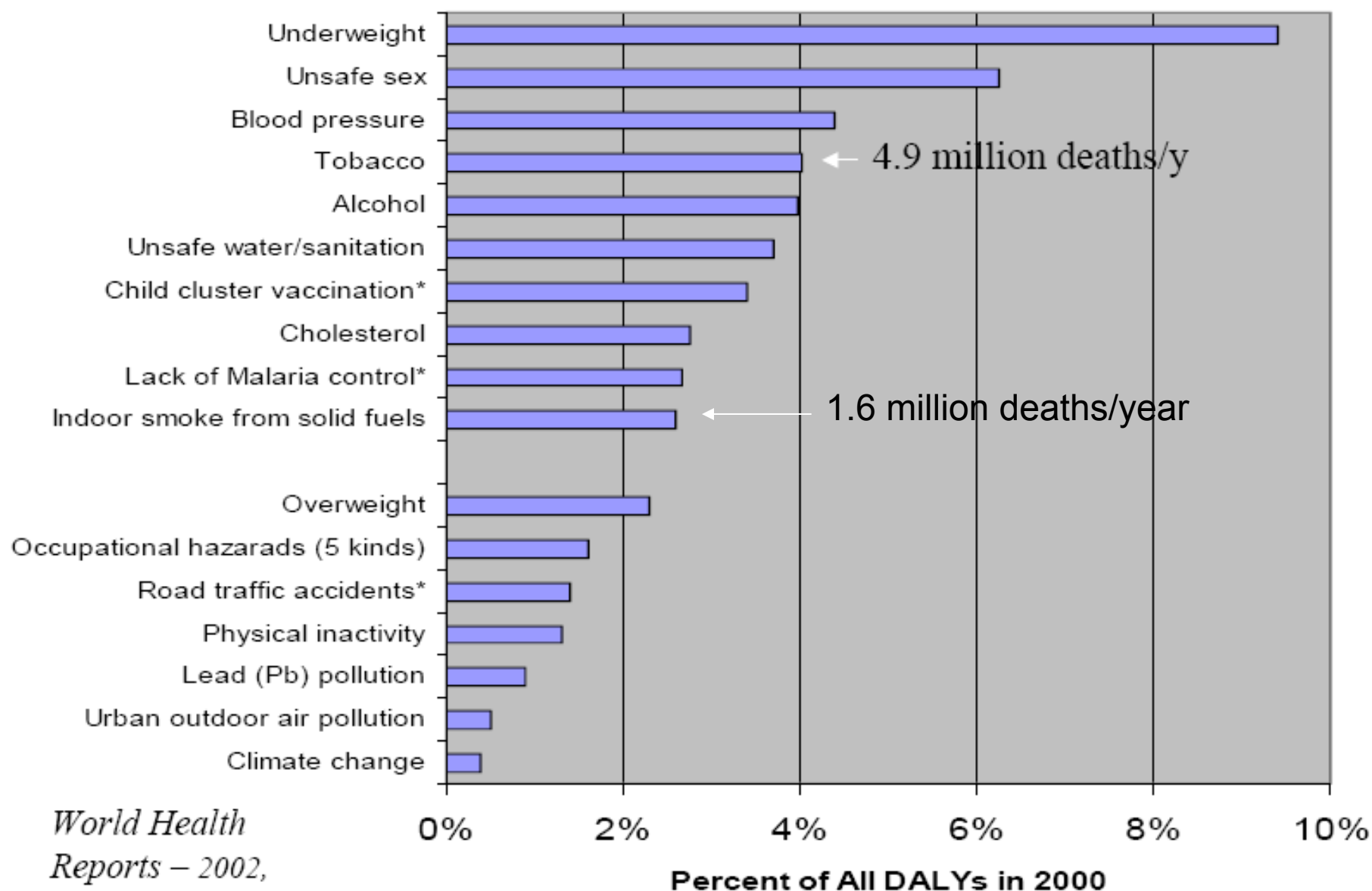
- 50% of the world's population relies on solid fuels
- Exposure to IAP from solid fuels is strongly associated with ARI, COPD, & lung cancer
- 1.5 to 2 million deaths annually

Source: Smith K, Mehta S, et al. (2004).
Comparative Quantification of Health Risks:
Global and Regional Burden of Disease due to
Selected Major Risk Factors.



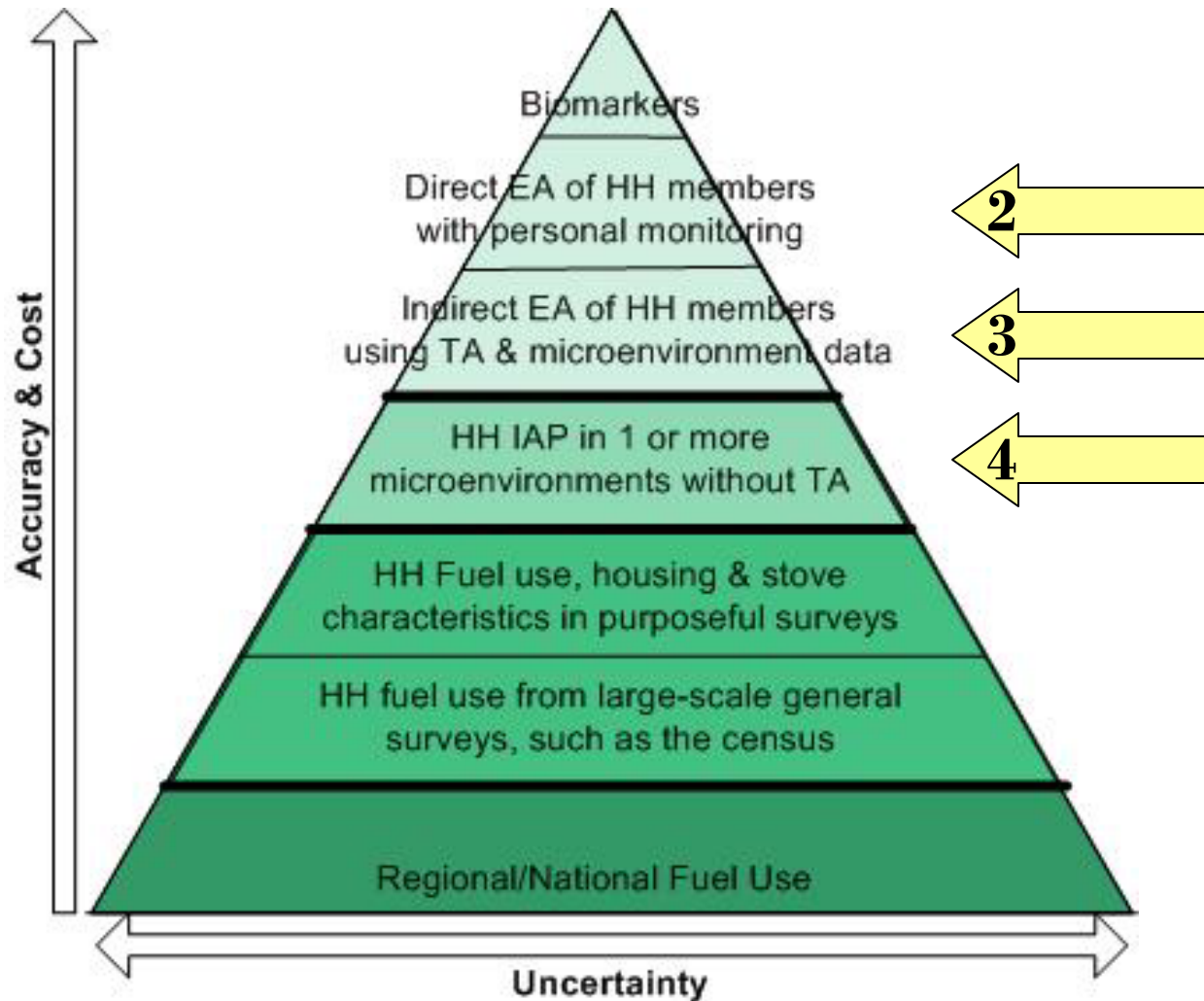
Chullah, traditional Indian stove in Uttarakhand, India

Global Burden of Disease from Top 10 Risk Factors plus selected other risk factors





The IAP Exposure Pyramid



Exposure Pyramid Research Aim

- Quantify the improvements, if any – in terms of accuracy – that result by shifting from indirect to direct methods in a 2-year CO exposure assessment of 64 rural women using wood fuel in highland Guatemala.

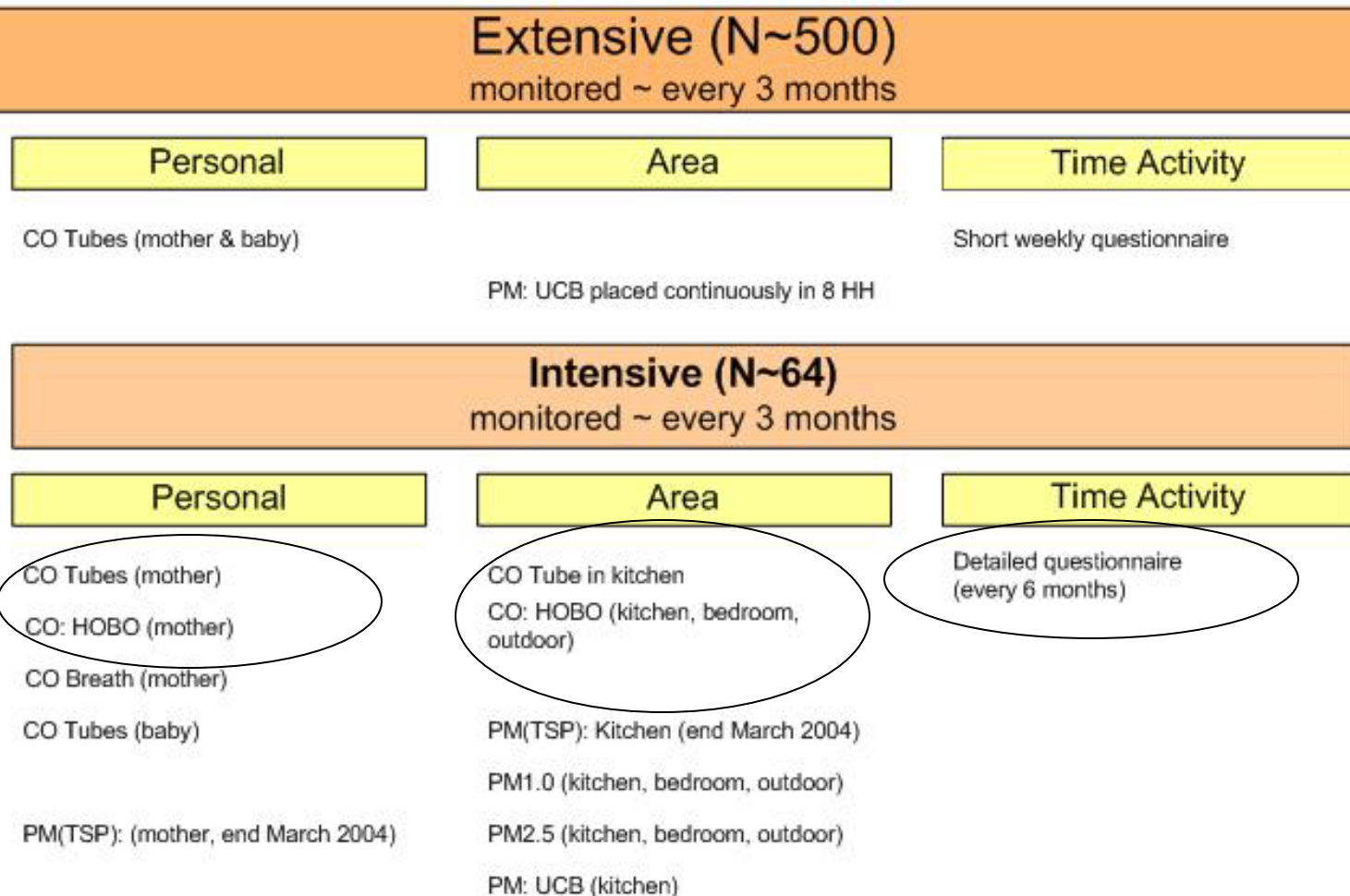
Study site (RESPIRE): Guatemala



plancha (improved stove)

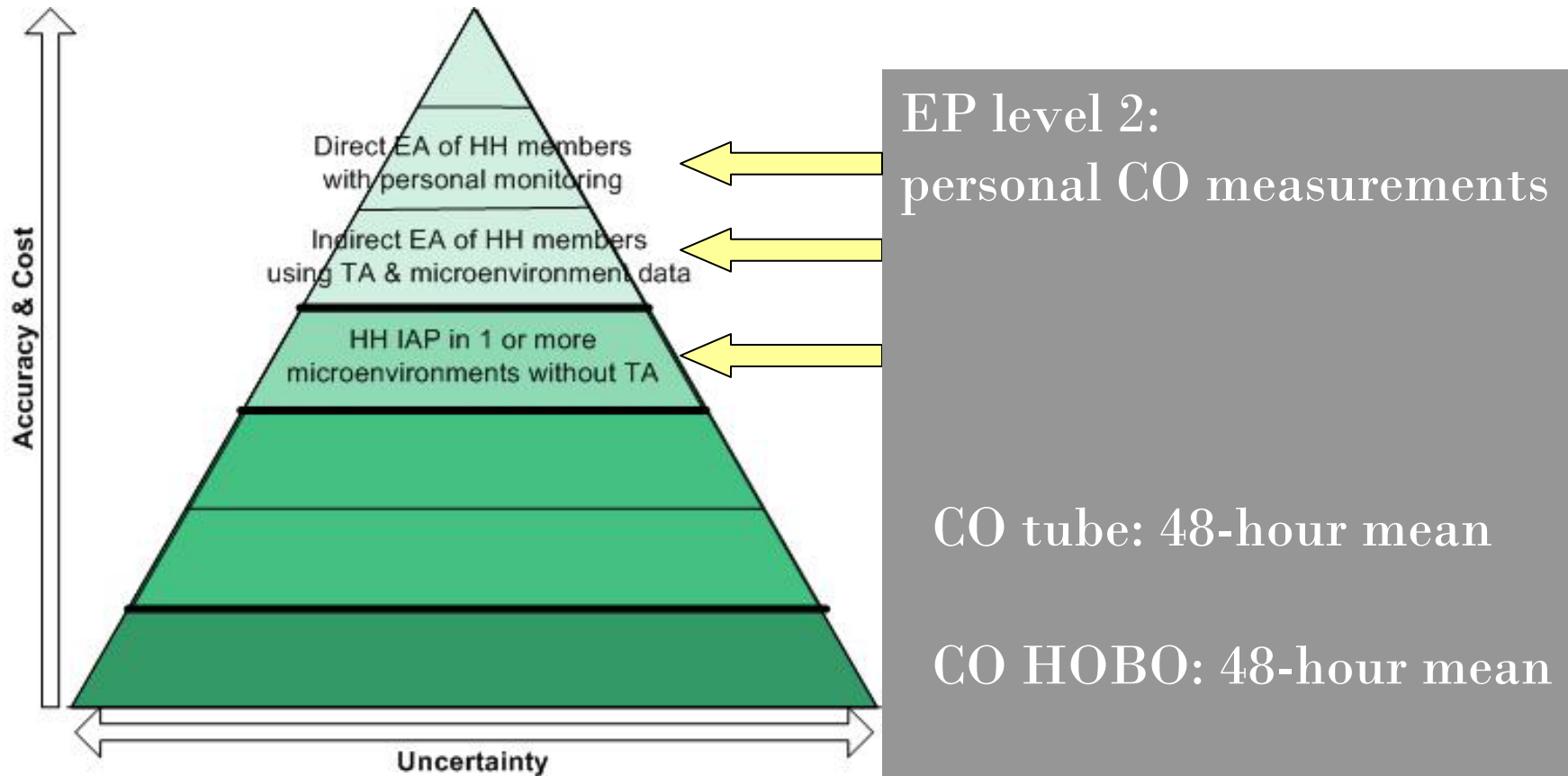
RESPIRE IAP data

Dec 2002 – Dec 2004



Calculation of 64 Women's CO Exposures

(data source: CO tube and CO HOBO)

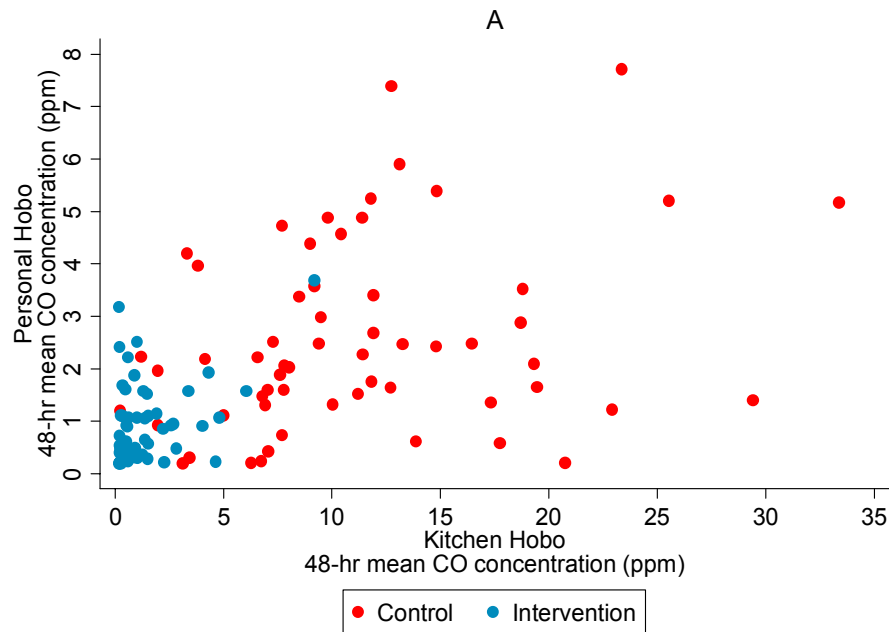


Methods to assess the relationship between metrics on each level of the Exposure Pyramid

- Least squares linear regression to calculate the coefficient of determination (R^2)
 - Spearman's correlation coefficient
 - Stepwise regression analysis to estimate personal CO measurements (including questionnaire data)
-

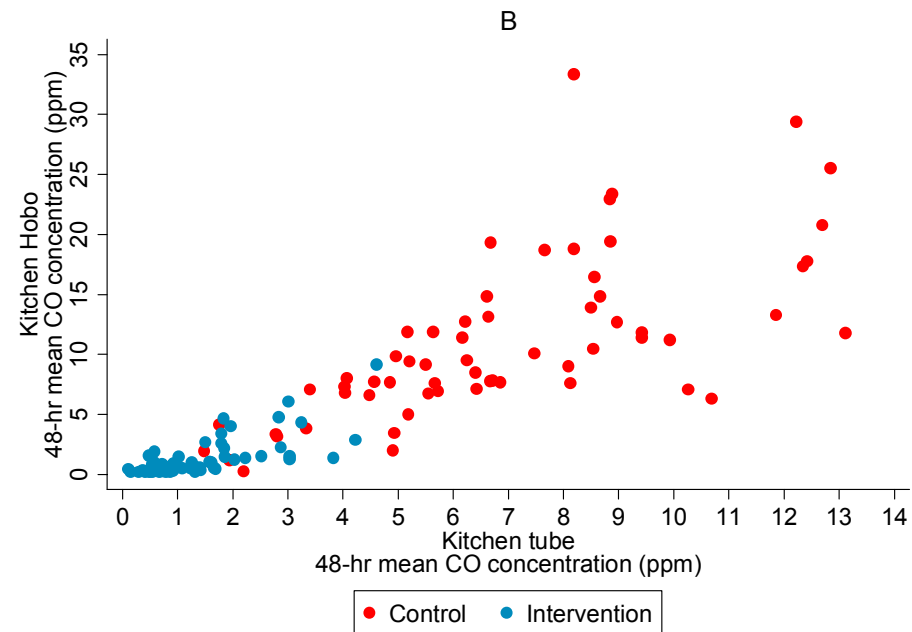
Correlations among Exposure Metrics at Different Levels of the Pyramid

Levels 2 and 4



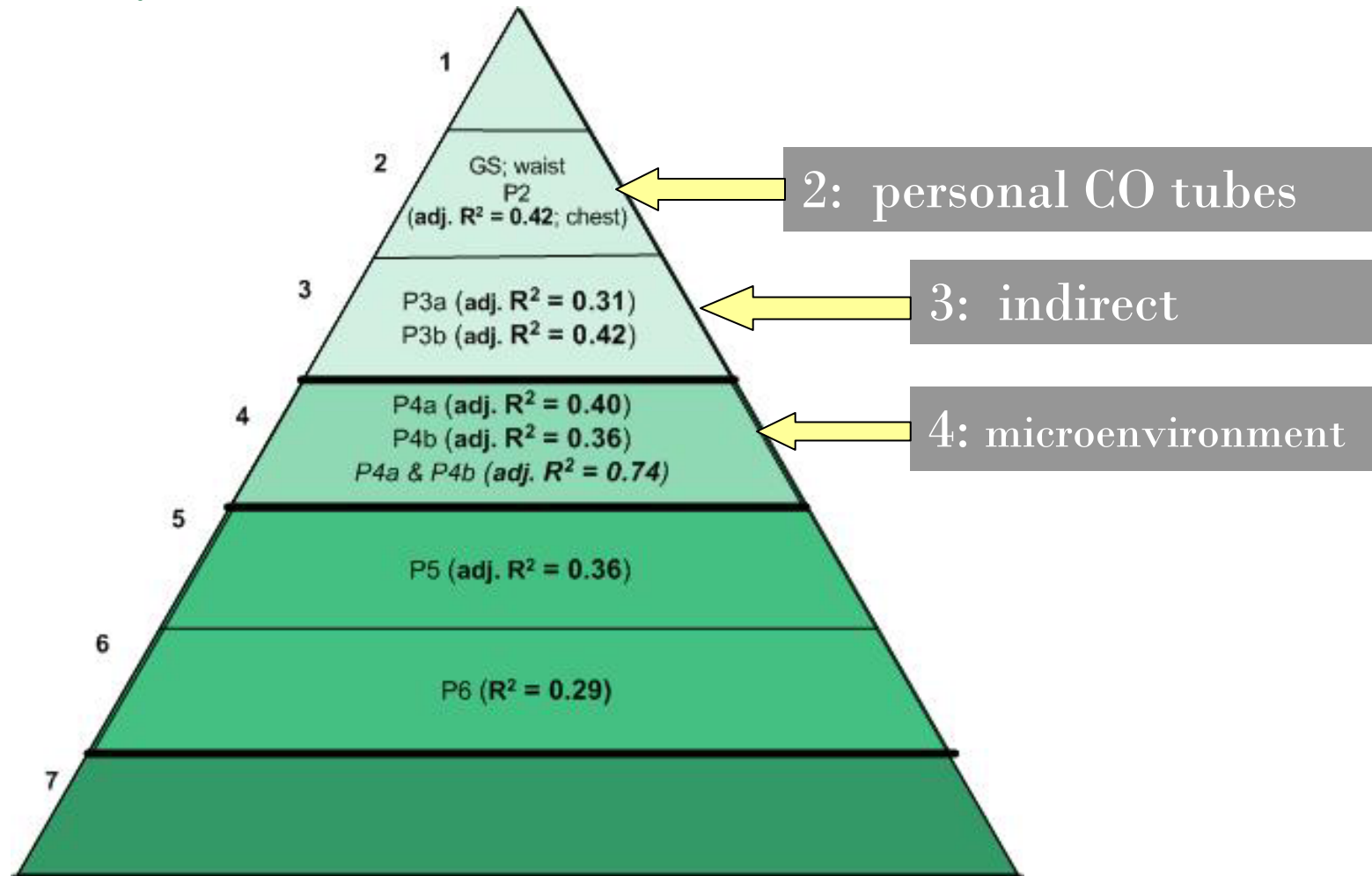
N=134
 $R^2 = 0.28$

Level 4



N=134
 $R^2 = 0.74$

Results: Correlation Strength Between each Proxy and the Gold Standard

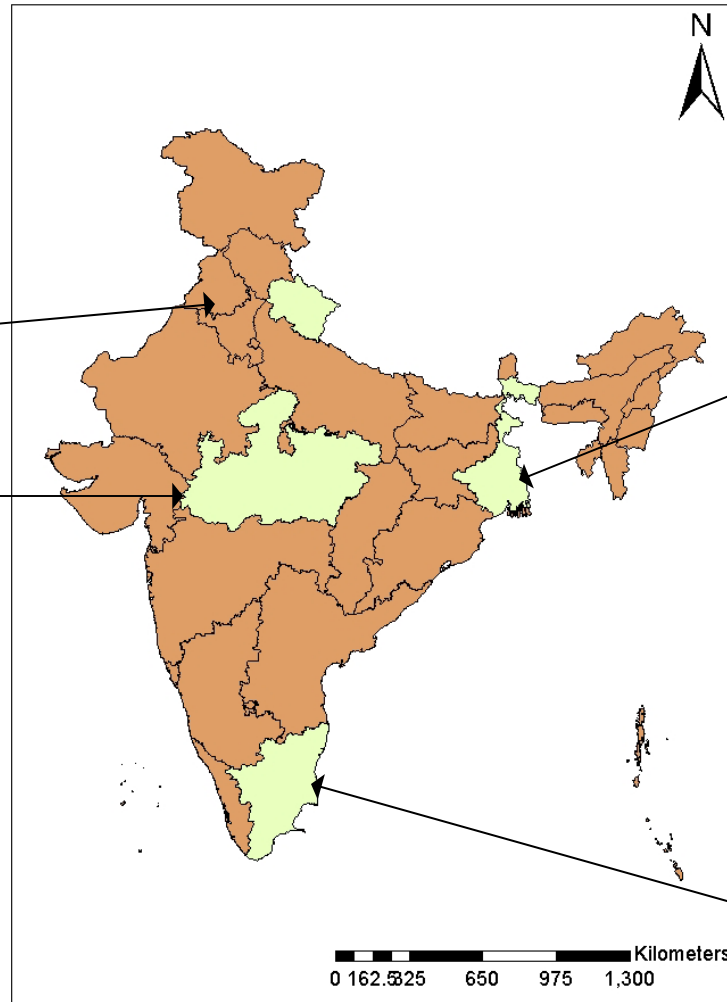


Results based on step-wise regression analysis.

Conclusions: Exposure Pyramid

- Correlation strength between the continuous personal CO measurements and simultaneously collected indirect and microenvironmental CO measurements ranged from adjusted $R^2 = 0.31 - 0.42$.
- Several factors may explain the lower than expected correlations, including instrument placement, inaccurate time activity data, HH ventilation patterns, and perhaps user error (results reported here are not atypical of similar results published in the literature).
- Analysis provides further justification for using area measurements to estimate exposure in studies by non-research groups.

Case study: Cross-sectional study of exposure to IAP and lung function in India

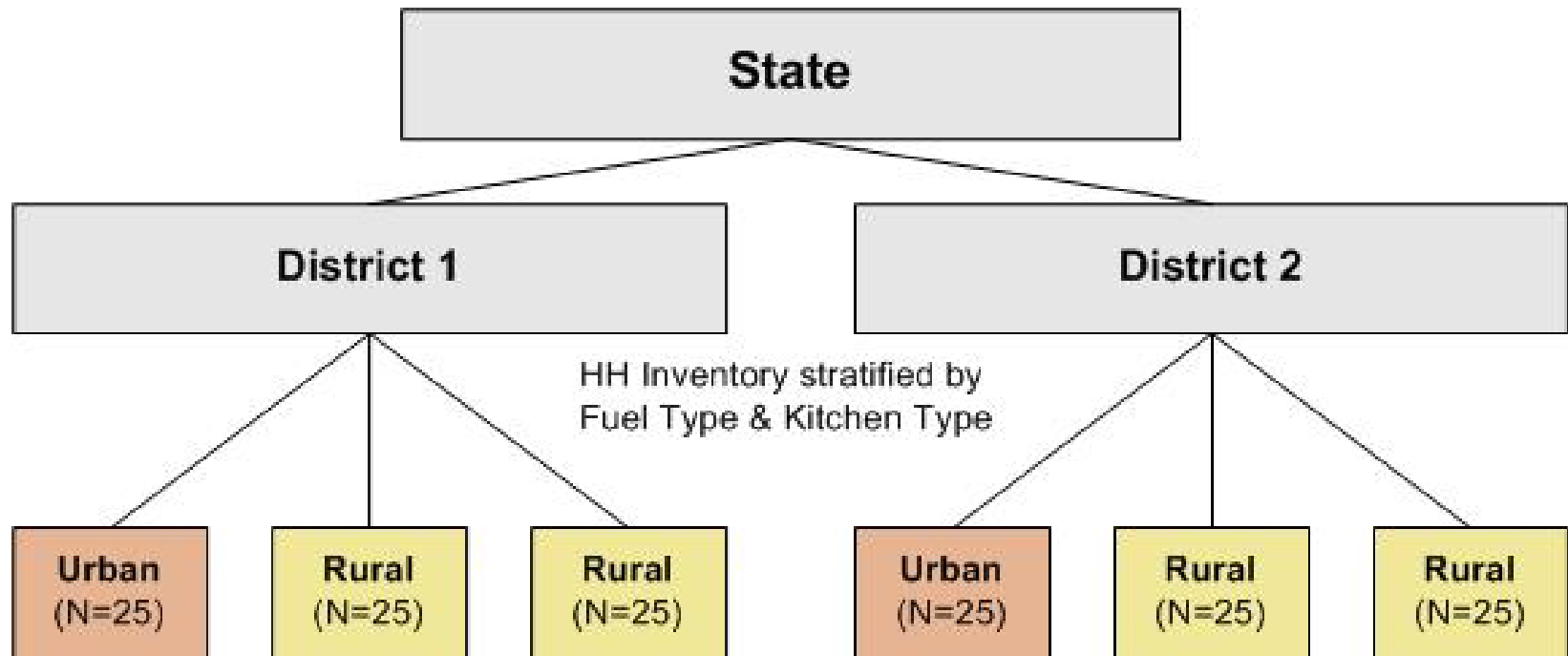








HEED Household Sampling Scheme



N = 150 households/state

N = 600 households total

HEED IAP data

(Nov 2004 – Feb 2005)

typical sampling day

HH1

Kitchen:
UCB & CO tube
Gravimetric
Living:
Grav & UCB
Outside:
some UCBs & Grav

HH2

Kitchen:
UCB & CO tube
Living:
UCB
Outside:
some UCBs

HH3

Kitchen:
UCB & CO tube
Living:
UCB
Outside:
some UCBs

HH4

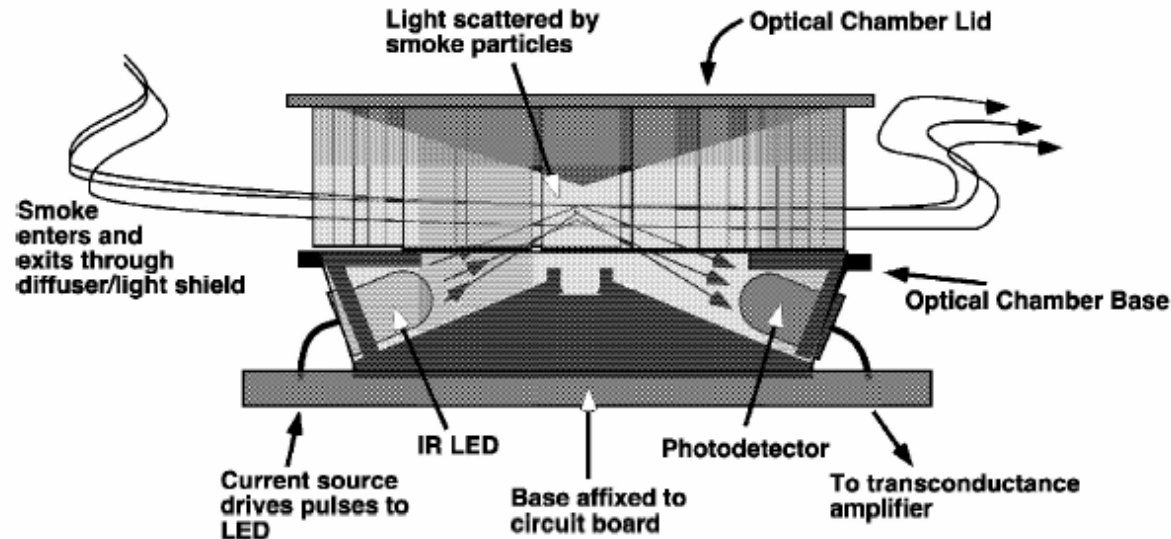
Kitchen:
UCB & CO tube
Living:
UCB
Outside:
some UCBs

HH5

Kitchen:
UCB & CO tube
Living:
UCB
Outside:
some UCBs

Respiratory questionnaires and spirometry tests completed in all households; physician examination in Tamil Nadu.

Measuring Particles: Standard gravimetric (PM_{2.5}) and UCB particle monitor



Sources

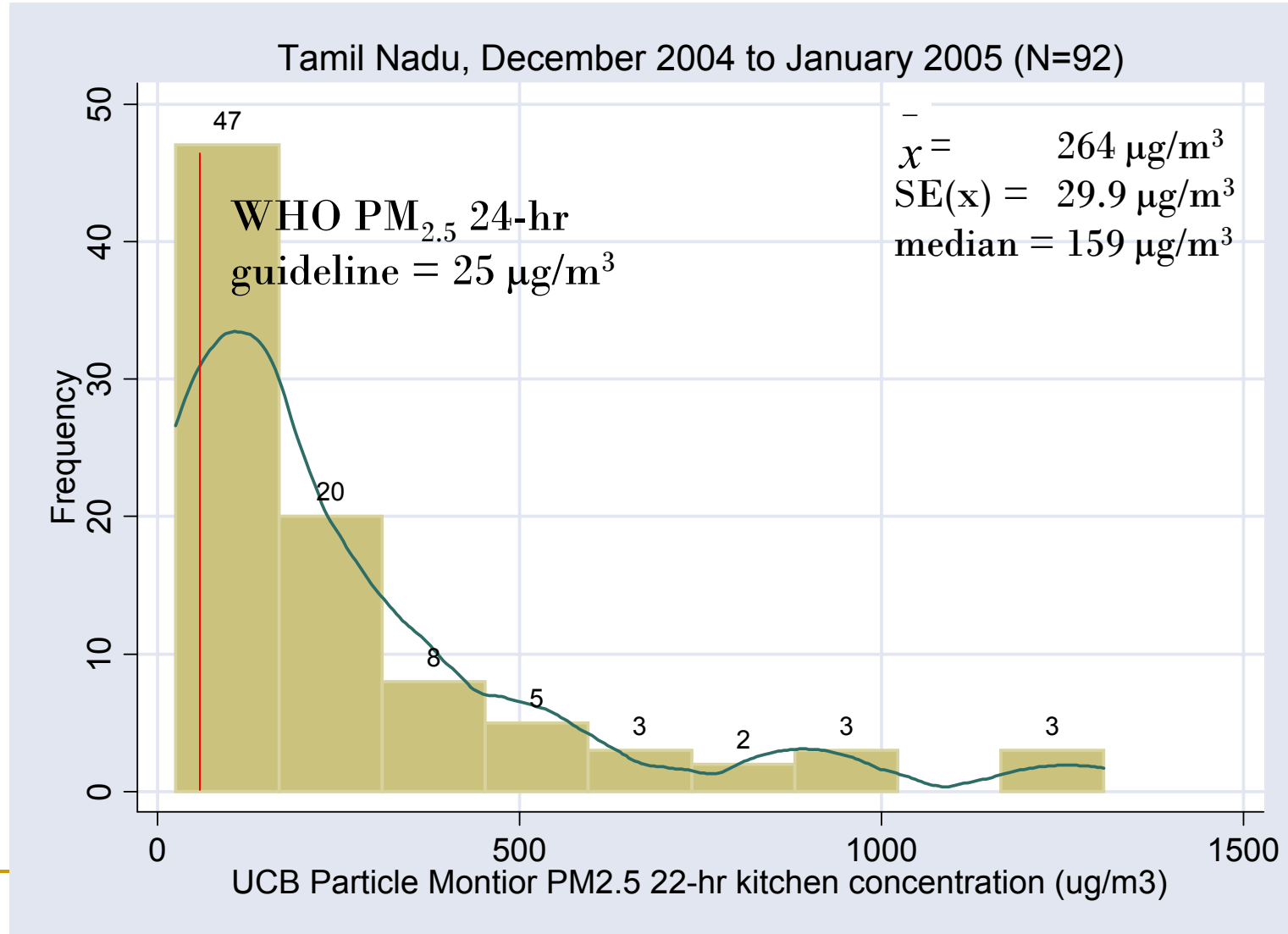
Chowdhury, Z., R. Edwards, et al. (2007). Journal of Environmental Monitoring [accepted 20 June 2007].

Edwards, R., K. R. Smith, et al. (2006). J Air Waste Manag Assoc 56(6): 789-99.

Litton, C. D., K. R. Smith, et al. (2004). Aerosol science and technology 38: 1054-1062.

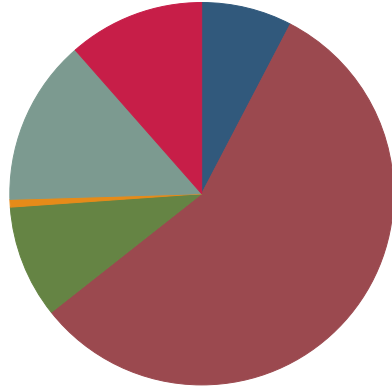
Typical Kitchen Concentrations in India

(all fuel/stove combinations)

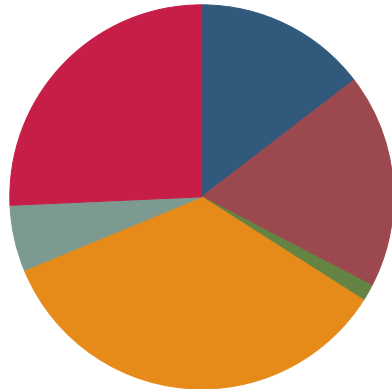


Results, 1: Differences in housing types by state

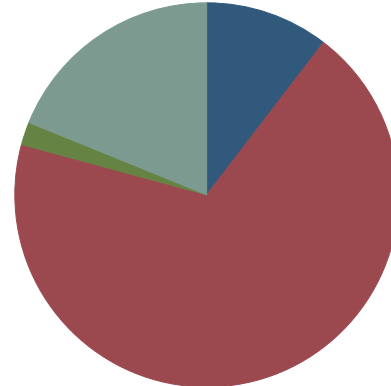
Housing type, Uttarranchal



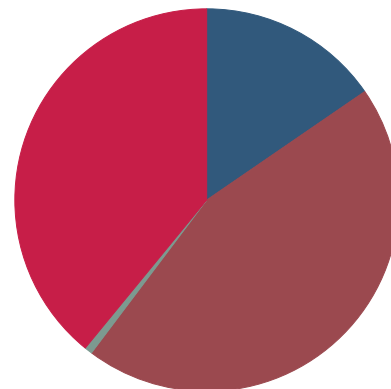
Housing type, Tamil Nadu



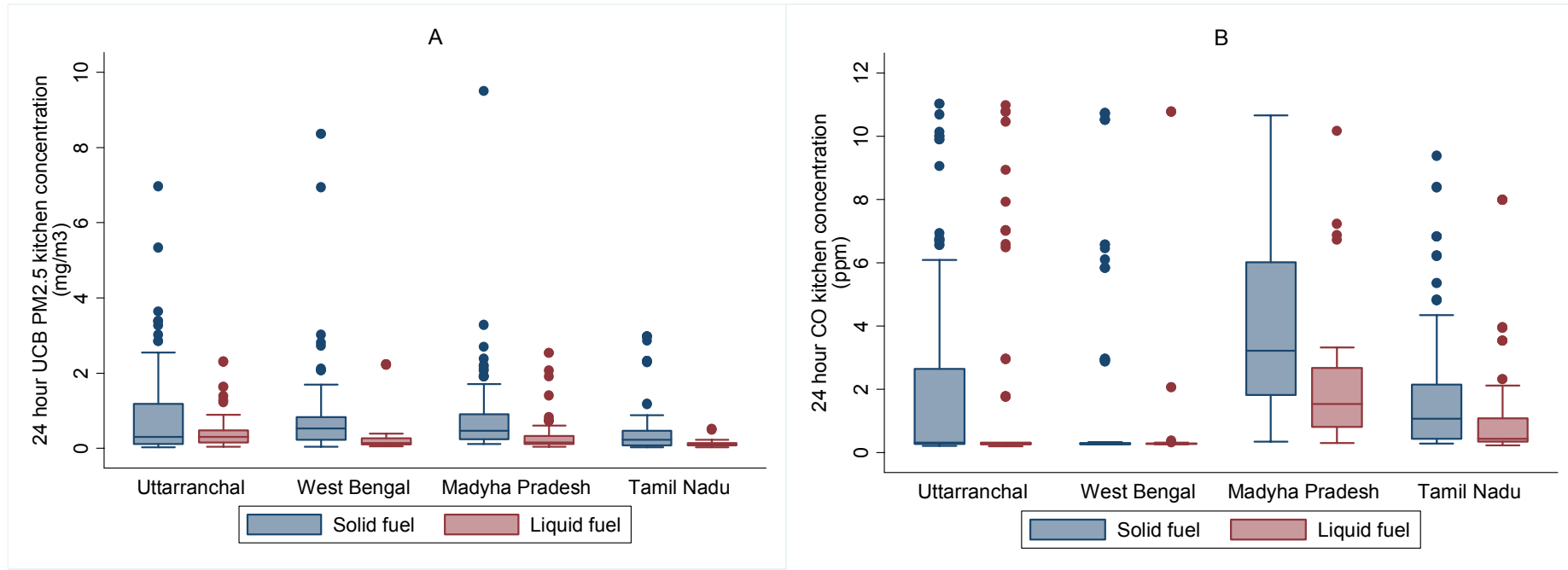
Housing type, Madyha Pradesh



Housing type, West Bengal



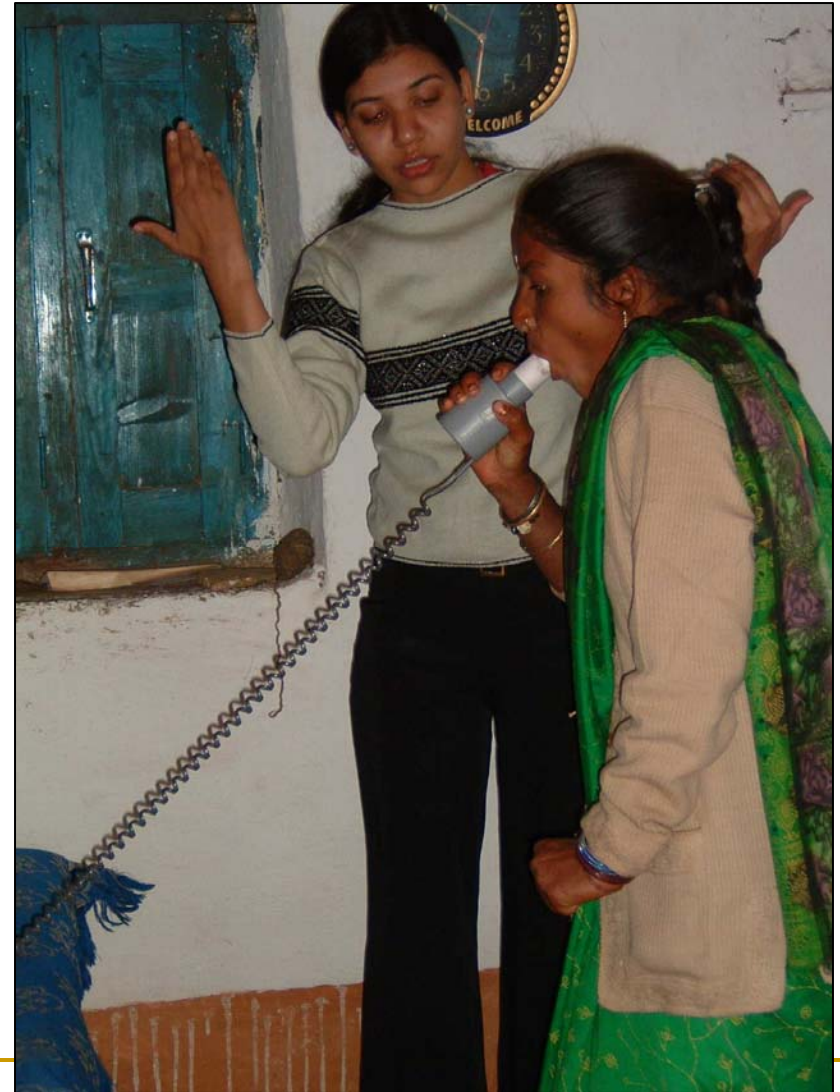
Results, 2: 24-hour PM and CO Concentrations



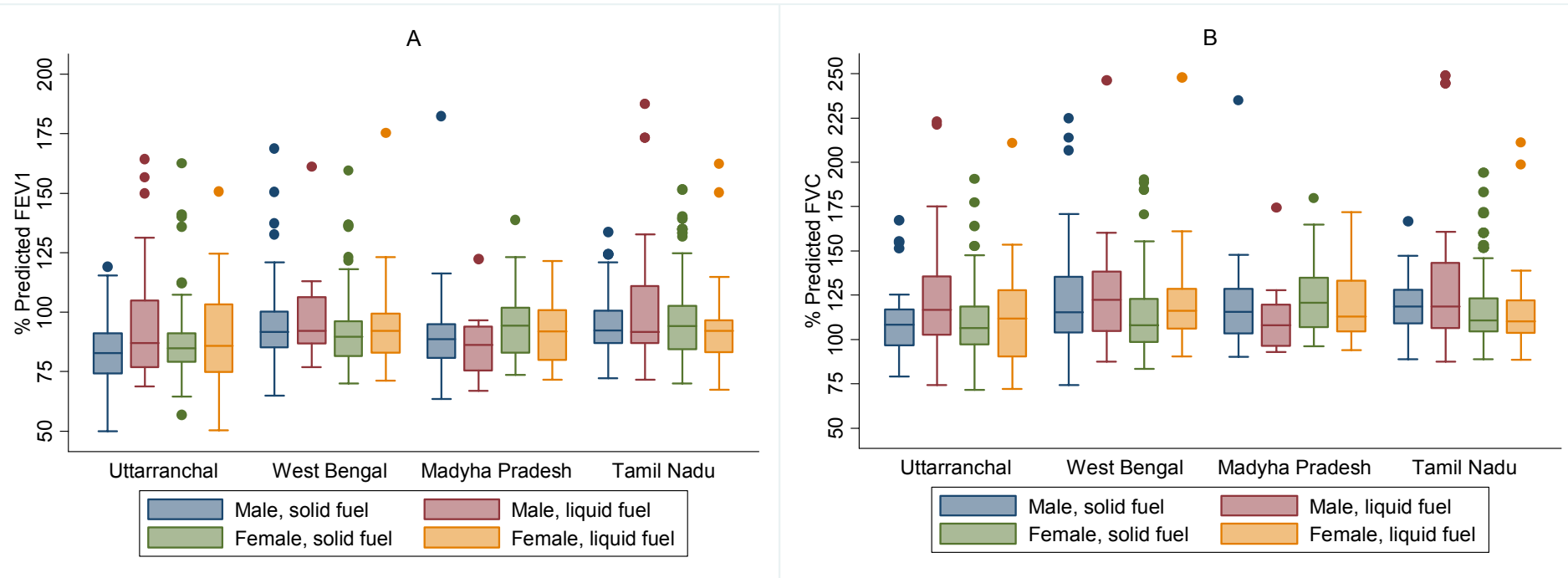
In all states, the mean 24-hour **PM_{2.5} concentration** in clean-fuel-using HH was significantly lower than that in solid-fuel-using HH.

In MP and TN, the mean 24-hour **CO concentration** in clean-fuel-using HH was significantly lower than that in solid-fuel-using HH.

Spirometry tests in households in India



Results, 3: % predicted FEV₁ and FVC values by sex in solid- and clean-fuel-using households



Among males, the mean % predicted FEV₁ and FVC were significantly higher among clean as compared to solid fuel users in Tamil Nadu.

The % predicted FEV₁ and FVC values differed significantly between males across the four states who lived in clean-fuel-using HH.

Conclusions: Case Study

- 24-hour mean kitchen concentration was $940 \mu\text{g}/\text{m}^3$ in solid fuel households.
- 24-hour mean kitchen concentration was $290 \mu\text{g}/\text{m}^3$ in clean fuel households.
- Values are 38 times and 12 times greater than the 2006 WHO Air Quality Guideline 24-hour mean $\text{PM}_{2.5}$ concentration of $25 \mu\text{g}/\text{m}^3$.
- The strongest effects on lung function (FEV_1) in **women** were detected relative to peak concentrations of PM, i.e. age and height adjusted 24-hour 1-minute maximum $\text{PM}_{2.5}$ concentration ($p=0.007$) and adjusted 24-hour maximum 15 minute mean $\text{PM}_{2.5}$ concentration ($p=0.042$).

Implications and Research Plans

- Findings continue to support advocacy for the implementation, use and assessment of improved cookstoves as well as for better household ventilation.
- Currently investigating an approach for estimating reductions in ALRI and COPD mortality based on IAP measurements.
- Research plans:
 - Replicate the Exposure Pyramid analysis in a more controlled-setting where ventilation conditions and other factors can be more closely controlled.
 - Evaluate how other metrics – including using spatial analyses such as krigging and linear interpolation to create IAP “exposure surfaces” – fit into the Exposure Pyramid framework.

Acknowledgements

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